

Creating Linear Regression Model Using PySpark

Roll No. Is : DS5B-2118

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```
In [ ]: pip install pyspark
```

```
Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simple/
Collecting pyspark
  Downloading pyspark-3.2.1.tar.gz (281.4 MB)
    |████████████████████████████████████████| 281.4 MB 34 kB/s
Collecting py4j==0.10.9.3
  Downloading py4j-0.10.9.3-py2.py3-none-any.whl (198 kB)
    |████████████████████████████████████████| 198 kB 46.5 MB/s
Building wheels for collected packages: pyspark
  Building wheel for pyspark (setup.py) ... done
  Created wheel for pyspark: filename=pyspark-3.2.1-py2.py3-none-any.whl size=281853642 sha256=99657e37a6edb52a83d4b4e280e11c4e26947120a4b2dc8192749712f371238e
  Stored in directory: /root/.cache/pip/wheels/9f/f5/07/7cd8017084dce4e93e84e92efd1e1d5334db05f2e83bcef74f
Successfully built pyspark
Installing collected packages: py4j, pyspark
Successfully installed py4j-0.10.9.3 pyspark-3.2.1
```

```
In [ ]: from pyspark.sql import SparkSession
```

```
In [ ]: session = SparkSession.builder.appName("exam1").master("local").getOrCreate()
```

Read Dataset

```
In [ ]: data = session.read.csv("Big Mart Sale.csv", header = True, inferSchema=True)
```

To print top 10 raw in dataset

```
In [ ]: data.show(10)
```

```
+-----+-----+-----+-----+-----+-----+-----+
|Item_Identifier|Item_Weight|Item_Fat_Content|Item_Visibility|Item_Type|Item_MRP|Outlet_Identifier|Outlet_Establishment_Year|Outlet_Size|Outlet_Location_Type|Outlet_Type|Item_Outlet_Sales|
+-----+-----+-----+-----+-----+-----+-----+
|FDA15|9.3|Low Fat|0.016047301|Dairy|249.8092|OUT049|1999|Medium|Tier 1|Supermarket|
```

```

rket Type1|      3735.138|
|          DRC01|      5.92|      Regular|      0.019278216|      Soft Drinks| 48.2
692|          OUT018|      2009|      Medium|      Tier 3|Superma
rket Type2|      443.4228|
|          FDN15|      17.5|      Low Fat|      0.016760075|      Meat| 141.
618|          OUT049|      1999|      Medium|      Tier 1|Superma
rket Type1|      2097.27|
|          FDX07|      19.2|      Regular|      0.0|Fruits and Vegeta...| 182.
095|          OUT010|      1998|      null|      Tier 3|      Gro
cery Store|      732.38|
|          NCD19|      8.93|      Low Fat|      0.0|      Household| 53.8
614|          OUT013|      1987|      High|      Tier 3|Superma
rket Type1|      994.7052|
|          FDP36|      10.395|      Regular|      0.0|      Baking Goods| 51.4
008|          OUT018|      2009|      Medium|      Tier 3|Superma
rket Type2|      556.6088|
|          FDO10|      13.65|      Regular|      0.012741089|      Snack Foods| 57.6
588|          OUT013|      1987|      High|      Tier 3|Superma
rket Type1|      343.5528|
|          FDP10|      null|      Low Fat|      0.127469857|      Snack Foods|107.7
622|          OUT027|      1985|      Medium|      Tier 3|Superma
rket Type3|      4022.7636|
|          FDH17|      16.2|      Regular|      0.016687114|      Frozen Foods| 96.9
726|          OUT045|      2002|      null|      Tier 2|Superma
rket Type1|      1076.5986|
|          FDU28|      19.2|      Regular|      0.09444959|      Frozen Foods|187.8
214|          OUT017|      2007|      null|      Tier 2|Superma
rket Type1|      4710.535|
+-----+-----+-----+-----+-----+-----+-----+
---+-----+-----+-----+-----+-----+-----+-----+
-----+-----+
only showing top 10 rows

```

Handle Null Values in columns

```
In [ ]: from pyspark.sql.functions import isnan, when, count, col
```

```
In [ ]: data.select([count(when(isnan(c) | col(c).isNull(), c)).alias(c) for c in data.columns])

+-----+-----+-----+-----+-----+-----+-----+
-----+-----+-----+-----+-----+-----+-----+
-----+
|Item_Identifier|Item_Weight|Item_Fat_Content|Item_Visibility|Item_Type|Item_MRP|Outlet_
Identifier|Outlet_Establishment_Year|Outlet_Size|Outlet_Location_Type|Outlet_Type|Item_O
utlet_Sales|
+-----+-----+-----+-----+-----+-----+-----+
-----+-----+-----+-----+-----+-----+-----+
-----+
|              0|              1463|              0|              0|              0|              0|
              0|              0|              2410|              0|              0|
              0|
+-----+-----+-----+-----+-----+-----+-----+
-----+-----+-----+-----+-----+-----+-----+
-----+

```

```
In [ ]: import pyspark.sql.functions as func
```

```
In [ ]: data.agg(func.percentile_approx("Item_Weight", 0.5).alias("mean")).show()
```

```

+----+
|mean|
+----+

```

```
|12.6|
+-----+
```

First we replace 12.6 in place of Null values in Item_weight column because it is mean in this column

```
In [ ]: data = data.na.fill(value=12.6,subset=["Item_Weight"])
```

Second we return Medium in place of Null values in Outlet_Size Column Because Medium is the median in Outlet_Size Column

```
In [ ]: data = data.na.fill(value="Medium",subset=["Outlet_Size"])
```

```
In [ ]: data.show()
```

```
+-----+-----+-----+-----+-----+-----+-----+
---+-----+-----+-----+-----+-----+-----+-----+
-----+-----+
|Item_Identifier|Item_Weight|Item_Fat_Content|Item_Visibility|Item_Type|Item_
MRP|Outlet_Identifier|Outlet_Establishment_Year|Outlet_Size|Outlet_Location_Type|O
utlet_Type|Item_Outlet_Sales|
+-----+-----+-----+-----+-----+-----+-----+
---+-----+-----+-----+-----+-----+-----+-----+
-----+-----+
|          FDA15|          9.3|          Low Fat|          0.016047301|          Dairy|249.8
092|          OUT049|          1999|          Medium|          Tier 1|Superma
rket Type1|          3735.138|
|          DRC01|          5.92|          Regular|          0.019278216|          Soft Drinks| 48.2
692|          OUT018|          2009|          Medium|          Tier 3|Superma
rket Type2|          443.4228|
|          FDN15|          17.5|          Low Fat|          0.016760075|          Meat| 141.
618|          OUT049|          1999|          Medium|          Tier 1|Superma
rket Type1|          2097.27|
|          FDX07|          19.2|          Regular|          0.0|Fruits and Vegeta...| 182.
095|          OUT010|          1998|          Medium|          Tier 3|Gro
cery Store|          732.38|
|          NCD19|          8.93|          Low Fat|          0.0|          Household| 53.8
614|          OUT013|          1987|          High|          Tier 3|Superma
rket Type1|          994.7052|
|          FDP36|          10.395|          Regular|          0.0|          Baking Goods| 51.4
008|          OUT018|          2009|          Medium|          Tier 3|Superma
rket Type2|          556.6088|
|          FDO10|          13.65|          Regular|          0.012741089|          Snack Foods| 57.6
588|          OUT013|          1987|          High|          Tier 3|Superma
rket Type1|          343.5528|
|          FDP10|          12.6|          Low Fat|          0.127469857|          Snack Foods|107.7
622|          OUT027|          1985|          Medium|          Tier 3|Superma
rket Type3|          4022.7636|
|          FDH17|          16.2|          Regular|          0.016687114|          Frozen Foods| 96.9
726|          OUT045|          2002|          Medium|          Tier 2|Superma
rket Type1|          1076.5986|
|          FDU28|          19.2|          Regular|          0.09444959|          Frozen Foods|187.8
214|          OUT017|          2007|          Medium|          Tier 2|Superma
rket Type1|          4710.535|
|          FDY07|          11.8|          Low Fat|          0.0|Fruits and Vegeta...| 45.5
402|          OUT049|          1999|          Medium|          Tier 1|Superma
rket Type1|          1516.0266|
|          FDA03|          18.5|          Regular|          0.045463773|          Dairy|144.1
102|          OUT046|          1997|          Small|          Tier 1|Superma
rket Type1|          2187.153|
|          FDX32|          15.1|          Regular|          0.1000135|Fruits and Vegeta...|145.4
786|          OUT049|          1999|          Medium|          Tier 1|Superma
rket Type1|          1589.2646|
|          FDS46|          17.6|          Regular|          0.047257328|          Snack Foods|119.6
```

```

782|          OUT046|          1997|          Small|          Tier 1|Superma
rket Type1|          2145.2076|
|          FDF32|          16.35|          Low Fat|          0.0680243|Fruits and Vegeta...|196.4
426|          OUT013|          1987|          High|          Tier 3|Superma
rket Type1|          1977.426|
|          FDP49|          9.0|          Regular|          0.069088961|          Breakfast| 56.3
614|          OUT046|          1997|          Small|          Tier 1|Superma
rket Type1|          1547.3192|
|          NCB42|          11.8|          Low Fat|          0.008596051|          Health and Hygiene|115.3
492|          OUT018|          2009|          Medium|          Tier 3|Superma
rket Type2|          1621.8888|
|          FDP49|          9.0|          Regular|          0.069196376|          Breakfast| 54.3
614|          OUT049|          1999|          Medium|          Tier 1|Superma
rket Type1|          718.3982|
|          DRI11|          12.6|          Low Fat|          0.034237682|          Hard Drinks|113.2
834|          OUT027|          1985|          Medium|          Tier 3|Superma
rket Type3|          2303.668|
|          FDU02|          13.35|          Low Fat|          0.10249212|          Dairy|230.5
352|          OUT035|          2004|          Small|          Tier 2|Superma
rket Type1|          2748.4224|
+-----+-----+-----+-----+-----+-----+-----+
---+-----+-----+-----+-----+-----+-----+
-----+-----+
only showing top 20 rows

```

EDA

To print all columns name

```
In [ ]: data.columns
```

```
Out[ ]: ['Item_Identifier',
        'Item_Weight',
        'Item_Fat_Content',
        'Item_Visibility',
        'Item_Type',
        'Item_MRP',
        'Outlet_Identifier',
        'Outlet_Establishment_Year',
        'Outlet_Size',
        'Outlet_Location_Type',
        'Outlet_Type',
        'Item_Outlet_Sales']
```

To count total numbers of raws in dataset

```
In [ ]: data.count()
```

```
Out[ ]: 8523
```

```
In [ ]: data.dtypes
```

```
Out[ ]: [('Item_Identifier', 'string'),
        ('Item_Weight', 'double'),
        ('Item_Fat_Content', 'string'),
        ('Item_Visibility', 'double'),
        ('Item_Type', 'string'),
        ('Item_MRP', 'double'),
        ('Outlet_Identifier', 'string'),
        ('Outlet_Establishment_Year', 'int'),
        ('Outlet_Size', 'string'),
        ('Outlet_Location_Type', 'string'),
```

```
('Outlet_Type', 'string'),  
('Item_Outlet_Sales', 'double')]
```

Data Preprocessing

VectorAssembler :- It is feature transformer that combine multiple columns into a single vector column.

StringIndexer :- It is use for mapping a string column to a index column that will be treated as a categorical column by spark.

OneHotEncoder :- It is an important technique for converting categorical attributes into a numeric vector

```
In [ ]: from pyspark.ml.feature import VectorAssembler, StringIndexer, OneHotEncoder  
  
In [ ]: str_index = StringIndexer(inputCols = ['Item_Identifier','Item_Fat_Content','Item_Type',  
  
In [ ]: one_hot = OneHotEncoder(inputCols =['Item_Identifier1','Item_Fat_Content1','Item_Type1',  
  
In [ ]: vector_ass = VectorAssembler(inputCols = ['Item_Weight','Item_Fat_Content2','Item_Visibi  
  
In [ ]: from pyspark.ml.regression import LinearRegression  
  
In [ ]: linear = LinearRegression(featuresCol="allfeatures", labelCol="Item_Outlet_Sales")
```

Create Pipeline for our Model

```
In [ ]: from pyspark.ml import Pipeline  
mypipeline = Pipeline(stages = [str_index, one_hot, vector_ass, linear])
```

Using randomsplit data is split into 78% of training and 22% of test as given

```
In [ ]: training, test = data.randomSplit([0.78, 0.22])
```

Model Training

```
In [ ]: lin_reg_model = mypipeline.fit(training)
```

Test our Model using testing data

```
In [ ]: result = lin_reg_model.transform(test)
```

```
In [ ]: result.show()
```

```
+-----+-----+-----+-----+-----+-----+-----+-----+  
-----+-----+-----+-----+-----+-----+-----+-----+  
-+-----+-----+-----+-----+-----+-----+-----+-----+  
-----+-----+-----+-----+-----+-----+-----+-----+  
---+-----+-----+-----+-----+-----+-----+-----+-----+  
-----+-----+-----+-----+-----+-----+-----+-----+  
|Item_Identifier|Item_Weight|Item_Fat_Content|Item_Visibility|  Item_Type|Item_MRP|Outle  
t_Identifier|Outlet_Establishment_Year|Outlet_Size|Outlet_Location_Type|      Outlet_Typ  
e|Item_Outlet_Sales|Item_Identifier1|Item_Fat_Content1|Item_Type1|Outlet_Identifier1|Out  
let_Establishment_Year1|Outlet_Size1|Outlet_Location_Type1|Outlet_Type1|  Item_Identifi  
er2|Item_Fat_Content2|      Item_Type2|Outlet_Identifier2|Outlet_Establishment_Year2| Outl  
et_Size2|Outlet_Location_Type2| Outlet_Type2|      allfeatures|      prediction|  
+-----+-----+-----+-----+-----+-----+-----+-----+
```

	DRA12	11.6	Low Fat	0.0	Soft Drinks	141.6154
	OUT045	2002	Medium		Tier 2	Supermarket Type
1	3829.0158	1051.0	0.0	8.0	7.0	
		7.0	0.0	1.0	0.0	(1553, [1051], [1.0])
	(4, [0], [1.0])	(15, [8], [1.0])	(9, [7], [1.0])		(8, [7], [1.0])	(2, [0], [1.0])
	(2, [1], [1.0])	(3, [0], [1.0])	(29, [0, 1, 14, 21, 22...])		2277.372227927723	
	DRA12	11.6	Low Fat	0.041112694	Soft Drinks	142.0154
	OUT018	2009	Medium		Tier 3	Supermarket Type
2	850.8924	1051.0	0.0	8.0	5.0	
		5.0	0.0	0.0	3.0	(1553, [1051], [1.0])
	(4, [0], [1.0])	(15, [8], [1.0])	(9, [5], [1.0])		(8, [5], [1.0])	(2, [0], [1.0])
	(2, [0], [1.0])	(3, [], [])	(29, [0, 1, 5, 14, 21, ...])		1937.56029983537	
	DRA12	11.6	Low Fat	0.068535039	Soft Drinks	143.0154
	OUT010	1998	Medium		Tier 3	Grocery Stor
e	283.6308	1051.0	0.0	8.0	8.0	
		8.0	0.0	0.0	1.0	(1553, [1051], [1.0])
	(4, [0], [1.0])	(15, [8], [1.0])	(9, [8], [1.0])		(8, [], [])	(2, [0], [1.0])
	(2, [0], [1.0])	(3, [1], [1.0])	(29, [0, 1, 5, 14, 21, ...])		327.38010298800816	
	DRA24	19.35	Regular	0.039920687	Soft Drinks	163.3868
	OUT035	2004	Small		Tier 2	Supermarket Type
1	3439.5228	322.0	1.0	8.0	1.0	
		2.0	1.0	1.0	0.0	(1553, [322], [1.0])
	(4, [1], [1.0])	(15, [8], [1.0])	(9, [1], [1.0])		(8, [2], [1.0])	(2, [1], [1.0])
	(2, [1], [1.0])	(3, [0], [1.0])	(29, [0, 2, 5, 14, 21, ...])		2671.7706482761328	
	DRA24	19.35	Regular	0.040154087	Soft Drinks	164.6868
	OUT017	2007	Medium		Tier 2	Supermarket Type
1	1146.5076	322.0	1.0	8.0	2.0	
		3.0	0.0	1.0	0.0	(1553, [322], [1.0])
	(4, [1], [1.0])	(15, [8], [1.0])	(9, [2], [1.0])		(8, [3], [1.0])	(2, [0], [1.0])
	(2, [1], [1.0])	(3, [0], [1.0])	(29, [0, 2, 5, 14, 21, ...])		2691.662378943857	
	DRA59	8.27	Regular	0.0	Soft Drinks	183.2924
	OUT017	2007	Medium		Tier 2	Supermarket Type
1	2406.2012	97.0	1.0	8.0	2.0	
		3.0	0.0	1.0	0.0	(1553, [97], [1.0])
	(4, [1], [1.0])	(15, [8], [1.0])	(9, [2], [1.0])		(8, [3], [1.0])	(2, [0], [1.0])
	(2, [1], [1.0])	(3, [0], [1.0])	(29, [0, 2, 14, 21, 22...])		3009.0974065421287	
	DRA59	12.6	Regular	0.127308434	Soft Drinks	186.6924
	OUT027	1985	Medium		Tier 3	Supermarket Type
3	7033.5112	97.0	1.0	8.0	4.0	
		0.0	0.0	0.0	2.0	(1553, [97], [1.0])
	(4, [1], [1.0])	(15, [8], [1.0])	(9, [4], [1.0])		(8, [0], [1.0])	(2, [0], [1.0])
	(2, [0], [1.0])	(3, [2], [1.0])	(29, [0, 2, 5, 14, 21, ...])		4393.2781578674	
	DRB01	7.39	Low Fat	0.082367244	Soft Drinks	187.753
	OUT049	1999	Medium		Tier 1	Supermarket Type
1	1518.024	1336.0	0.0	8.0	3.0	
		4.0	0.0	2.0	0.0	(1553, [1336], [1.0])
	(4, [0], [1.0])	(15, [8], [1.0])	(9, [3], [1.0])		(8, [4], [1.0])	(2, [0], [1.0])
	(2, [], [])	(3, [0], [1.0])	(29, [0, 1, 5, 14, 21, ...])		2996.614390466222	
	DRB13	6.115	Regular	0.007043008	Soft Drinks	190.353
	OUT035	2004	Small		Tier 2	Supermarket Type
1	569.259	1052.0	1.0	8.0	1.0	
		2.0	1.0	1.0	0.0	(1553, [1052], [1.0])
	(4, [1], [1.0])	(15, [8], [1.0])	(9, [1], [1.0])		(8, [2], [1.0])	(2, [1], [1.0])
	(2, [1], [1.0])	(3, [0], [1.0])	(29, [0, 2, 5, 14, 21, ...])		3120.027487849694	
	DRB13	6.115	Regular	0.01179078	Soft Drinks	189.053
	OUT010	1998	Medium		Tier 3	Grocery Stor
e	948.765	1052.0	1.0	8.0	8.0	
		8.0	0.0	0.0	1.0	(1553, [1052], [1.0])
	(4, [1], [1.0])	(15, [8], [1.0])	(9, [8], [1.0])		(8, [], [])	(2, [0], [1.0])
	(2, [0], [1.0])	(3, [1], [1.0])	(29, [0, 2, 5, 14, 21, ...])		1146.3741600280819	
	DRB25	12.3	Low Fat	0.069446588	Soft Drinks	106.3938

1	OUT035	857.5504	2004	Small	Tier 2	Supermarket Type
0))	(4, [0], [1.0])	(15, [8], [1.0])	(9, [1], [1.0])	(8, [2], [1.0])	(2,	[1], [1.0])
	DRB48	12.6	Regular	0.024733134	Soft Drinks	40.2822
3	OUT027	1296.3126	1985	Medium	Tier 3	Supermarket Type
0))	(4, [1], [1.0])	(15, [8], [1.0])	(9, [4], [1.0])	(8, [0], [1.0])	(2,	[0], [1.0])
	DRB48	16.75	Regular	0.024848788	Soft Drinks	39.9822
1	OUT035	746.3618	2004	Small	Tier 2	Supermarket Type
0))	(4, [1], [1.0])	(15, [8], [1.0])	(9, [1], [1.0])	(8, [2], [1.0])	(2,	[1], [1.0])
	DRB48	16.75	Regular	0.041599644	Soft Drinks	40.9822
e	OUT010	157.1288	1998	Medium	Tier 3	Grocery Stor
0))	(4, [1], [1.0])	(15, [8], [1.0])	(9, [8], [1.0])	(8, [], [])	(2,	[0], [1.0])
	DRC01	5.92	Regular	0.019278216	Soft Drinks	48.2692
2	OUT018	443.4228	2009	Medium	Tier 3	Supermarket Type
0))	(4, [1], [1.0])	(15, [8], [1.0])	(9, [5], [1.0])	(8, [5], [1.0])	(2,	[0], [1.0])
	DRC01	5.92	Regular	0.019308607	Soft Drinks	49.0692
1	OUT017	1478.076	2007	Medium	Tier 2	Supermarket Type
0))	(4, [1], [1.0])	(15, [8], [1.0])	(9, [2], [1.0])	(8, [3], [1.0])	(2,	[0], [1.0])
	DRC12	17.85	Low Fat	0.03781972	Soft Drinks	191.6188
1	OUT035	2475.4444	2004	Small	Tier 2	Supermarket Type
0))	(4, [0], [1.0])	(15, [8], [1.0])	(9, [1], [1.0])	(8, [2], [1.0])	(2,	[1], [1.0])
	DRC12	17.85	Low Fat	0.037826873	Soft Drinks	189.7188
1	OUT046	2285.0256	1997	Small	Tier 1	Supermarket Type
0))	(4, [0], [1.0])	(15, [8], [1.0])	(9, [6], [1.0])	(8, [6], [1.0])	(2,	[1], [1.0])
	DRC12	17.85	Low Fat	0.038040837	Soft Drinks	189.1188
1	OUT017	3237.1196	2007	Medium	Tier 2	Supermarket Type
0))	(4, [0], [1.0])	(15, [8], [1.0])	(9, [2], [1.0])	(8, [3], [1.0])	(2,	[0], [1.0])
	DRC13	8.26	Regular	0.032573725	Soft Drinks	125.073
2	OUT018	985.384	2009	Medium	Tier 3	Supermarket Type
0))	(4, [1], [1.0])	(15, [8], [1.0])	(9, [5], [1.0])	(8, [5], [1.0])	(2,	[0], [1.0])
+	-----+	-----+	-----+	-----+	-----+	-----+
+	-----+	-----+	-----+	-----+	-----+	-----+
+	-----+	-----+	-----+	-----+	-----+	-----+
+	-----+	-----+	-----+	-----+	-----+	-----+
+	-----+	-----+	-----+	-----+	-----+	-----+

only showing top 20 rows

Evaluate Model accuracy

```
In [ ]: from pyspark.ml.evaluation import RegressionEvaluator
```

```
In [ ]: errors = ["r2", "rmse", "mse", "mae"]
name = ["R-Square or Accuracy", "Root Mean Square Error", "Mean Square Error", "Mean Abs

for i in range(len(errors)):
    eval = RegressionEvaluator(predictionCol="prediction", labelCol='Item_Outlet_Sales', m
    print("The {} of Model is {}".format(name[i],eval.evaluate(result)))
```

```
The R-Square or Accuracy of Model is 0.5609324399455548
The Root Mean Square Error of Model is 1146.154277794764
The Mean Square Error of Model is 1313669.6285072374
The Mean Absolute Error of Model is 854.720185337692
```

```
In [ ]:
```